

# CS231n Project Design

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# Agenda

1. Project expectations
  - a. Does my project meet expectations?
  - b. FAQs
2. Picking a project idea
  - a. Sources of inspiration
  - b. Reading papers efficiently
3. Proposal, milestone, and final report
  - a. Due dates, expectations, logistics
  - b. Support

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# Project expectations

The course project is a (fun) way to explore concepts taught in the course on a topic of your choice!

- Fairly open-ended, anything related to vision ([link to project page](#))

Completed in groups of 1, 2, or 3 people

- Project expectations are higher for groups with more people

Generally, two tracks of work:

- **Applications:** If you have a specific background or interest (e.g. biology, engineering, physics), we'd love to see you apply ConvNets to problems related to your particular domain of interest.
- **Models:** You can build a new model (algorithm) and apply it to tackle vision tasks. This track might be more challenging, and may later lead to a piece of publishable work.

# Project expectations

The final report has the following structure:

- Title, Author(s)
- Abstract
- Related Work
- Data Description
- Methods
- Experiments
- Conclusion
- Supplementary Material (optional)

For more on each of these, please also refer to the "final report" part of the **project page**

(<http://cs231n.stanford.edu/project.html#report>)

# FAQ: Does my project meet expectations?

Rule of thumb:

***- How much effort are you putting into your project?***

Strong projects might...

- Propose a novel variant of a technique (which takes a lot of effort)
- Adapt an existing technique to a totally new problem (which takes a lot of effort)

Weaker projects might...

- Spend several weeks collecting/cleaning data rather than testing hypotheses
- Clone an existing repo and do minimal stitching to make it work for a Kaggle competition

# FAQ: Does my project meet expectations?

So, this **doesn't** mean:

- Your project has to be strictly novel to get a good grade (although, we encourage this!)
- You have to beat the state-of-the-art performance to get a good grade (you don't have to come up with the next best object detector to test an interesting hypothesis)

This **does** mean:

- You need to put a significant effort into your investigation, and you may have to try many different approaches

In your **analysis**, ask yourself:

- Are you *interpreting* and *understanding* your results, or merely stating them?
- Are you just plotting a loss curve, or are you evaluating the results of your approach from many different angles?

# Project FAQs

**Q:** Can I apply convolutional networks to a purely NLP / audio / stock price problem?

- **A:** This is a computer vision course, so ***you must incorporate visual data in some form.***

**Q:** Can I change my project after the proposal, before the milestone?

- **A:** Yes - the proposal is to make sure you have a plausible project direction. If you need to change project directions, we understand.

**Q:** Can I change my project ***after*** the milestone?

- **A:** In general, we do not encourage this. At this point in the course, there will be little time to put together a sufficient project.



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# Picking a project idea

First and foremost:

**Do what is important or interesting to you, *not* what seems easiest.**

- You will be far more motivated if you're invested in what you're doing
- What do you *really* care about? Healthcare? Self-driving cars? Surveillance? Sports? Ethics? You can probably find its intersection with computer vision

Practical considerations:

1. **Data:** Is there existing data for this problem? Will I need to spend weeks collecting it myself?
2. **Code & framework:** Will I have to implement this myself, or is there an existing implementation?

# Picking a project idea

## Conferences:

[CVPR](#): IEEE Conference on Computer Vision and Pattern Recognition

[ICCV](#): International Conference on Computer Vision

[ECCV](#): European Conference on Computer Vision

[NeurIPS](#): Neural Information Processing Systems

[ICLR](#): International Conference on Learning Representations

[ICML](#): International Conference on Machine Learning

*Note: Do **not** even begin to try to read through all of these papers, or even their titles. There are far too many. Use CMD+F to find papers with relevant keywords.*

# Picking a project idea

## Additional resources:

- [Stanford Vision Lab Publications](#)
- [Awesome Deep Vision](#)
- [Papers With Code](#)
- [Kaggle](#)
- [Previous CS229 Projects](#)

# Reading papers

Do **not** read a paper linearly on your first pass

- First, read the abstract (word for word) as well as the figures & captions
- Does the paper still seem relevant? If so, read the methods as well as the results
- Finally, read the entire paper linearly (if the additional detail seems useful)

Papers are not always the most efficient way to digest an idea. Also try looking around

for:

- Talks, videos, or blog posts on the topics
- Github repos, containing actual code for the idea

# Reading papers

Example:

## **You Only Look Once: Unified, Real-Time Object Detection**

Joseph Redmon\*, Santosh Divvala\*<sup>†</sup>, Ross Girshick<sup>¶</sup>, Ali Farhadi\*<sup>†</sup>

University of Washington\*, Allen Institute for AI<sup>†</sup>, Facebook AI Research<sup>¶</sup>

<http://pjreddie.com/yolo/>

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# Deliverables

Due dates:

- Proposal (4/18) - Monday!
- Milestone (5/7)
- Final report (6/2)
- Poster session (6/4)



# Project Proposal (Due 4/18)

The proposal is a 200-400 word paragraph answering the following questions:

- **Problem:** What is the problem that you will be investigating? Why is it interesting?
- **Background:** What reading will you examine to provide context/background?
- **Data:** What data will you use?
  - If you are collecting new data, how will you do it?
- **Method:** What method or algorithm are you proposing?
  - If there are existing implementations, will you use them and how?
  - How do you plan to improve or modify such implementations?
- **Results:** How will you evaluate your results?
  - Qualitatively, what kind of results do you expect (e.g. plots or figures)?
  - Quantitatively, what kind of metrics will you use to evaluate/compare your results?

# Milestone (Due 5/7)

At most 3-page progress report, more or less containing:

**1. Literature review (3+ sources)**

**2. Indication that code is up and running**

3. Data source explained correctly

4. What Github repo or other code you're basing your work off of

**5. Ran baseline model have results**

a. Yes, points are taken off for no model running & no preliminary results

6. Data pipeline should be in place

7. Brief discussion of your preliminary results

# Support: CA areas of specialty

Day	Staff			
Mon	<b>Drew</b> <i>GANs, NLP</i>	<b>Yinan</b> <i>GANs, 3D vision, inverse graphics</i>	<b>Zhuoyi</b> <i>Object detection, scene graph generation, activity parsing; health care</i>	
Tue	<b>Mihir</b> <i>Autonomy, multi-object tracking, 3D vision, robotics (social navigation)</i>	<b>Manasi</b> <i>Image classification, image augmentation, 3D vision; astrophysics</i>	<b>Stephen</b> <i>Video understanding, action recognition, medical imaging, multi-task learning; healthcare</i>	<b>Dhruva</b> <i>CV, RL, NLP</i>
Wed	<b>Agrim</b> <i>RL, videos, detection; robotics, autonomous vehicles</i>	<b>Manasi</b> <i>Image classification, image augmentation, 3D vision; astrophysics</i>	<b>William</b> <i>3D vision, robot learning</i>	<b>Gokul</b> <i>RL, robot learning, simulation; robotics, autonomous vehicles</i>
Thu	<b>Shyamal</b> <i>Videos, vision+language</i>	<b>Bohan</b> <i>Robot learning; robotics</i>	<b>Dhruva</b> <i>CV, RL, NLP</i>	<b>Yinan</b> <i>GANs, 3D vision, inverse graphics</i>
Fri	<b>Agrim</b> <i>RL, videos, detection; robotics, autonomous vehicles</i>	<b>Moo Jin</b> <i>Robot learning, RL; robotics</i>	<b>Stephen</b> <i>Video understanding, action recognition, medical imaging, multi-task learning; healthcare</i>	<b>Sumith</b> <i>Video ML, human motion (sports/exercises)</i>
Sat	<b>Hongyu</b> <i>Graph neural networks</i>	<b>Bohan</b> <i>Robot learning; robotics</i>		
Sun	<b>Haochen</b> <i>Robot learning; robotics</i>	<b>Bohan</b> <i>Robot learning; robotics</i>		

**Questions?**